# 5 Lecture 5

# 5.1 Matrix multiplication

**Lemma 5.1.**  $T, S: V \to W$  be linear transformations. Let  $\beta = (v_1, \ldots, v_n)$  be a basis of V and  $\gamma = (w_1, \ldots, w_m)$  be a basis of W. Then  $[T+S]^{\gamma}_{\beta} = [T]^{\gamma}_{\beta} + [S]^{\gamma}_{\beta}$  and  $[cT]^{\gamma}_{\beta} = c[T]^{\gamma}_{\beta}$ .

*Proof.* Use linearity of  $T, S, \phi_{\beta}, \phi_{\gamma}$  to verify.

Corollary 5.2.  $\mathcal{L}(V, W) \cong M_{m \times n}(F)$  under the map  $T \mapsto [T]_{\beta}^{\gamma}$ . In particular, dim  $\mathcal{L}(V, W) = (\dim V)(\dim W)$ .

*Proof.* Show injectivity and surjectivity of the map  $T \mapsto [T]^{\gamma}_{\beta}$  using  $\beta$  and  $\gamma$  are bases.  $\Box$ 

**Definition 5.3.** Let  $A \in M_{m \times n}(F)$  and  $B \in M_{n \times p}(F)$ . Then the **matrix multiplication**  $C = AB \in M_{m \times p}(F)$  is given by

$$c_{ij} = \sum_{k=1}^{n} a_{ik} b_{kj}.$$

The *i*-th row, *j*-th column of C is given by multiplying the *i*-th row of A and *j*-th column of B.

**Proposition 5.4** (Matrix multiplication is composition of linear transformation). Let  $T: V \to W$  and  $S: W \to U$  be linear transformations, and  $\beta$ ,  $\gamma$  and  $\delta$  are bases of V, W, U respectively. Then

$$[ST]^{\delta}_{\beta} = [S]^{\delta}_{\gamma} [T]^{\gamma}_{\beta}.$$

*Proof.* We write  $[S]_{\gamma}^{\delta} = (a_{ij}) \in M_{m \times n}(F)$  and  $[T]_{\beta}^{\gamma} = (b_{ij}) \in M_{n \times p}(F)$  and  $\beta = (v_1, \ldots, v_p), \ \gamma = (w_1, \ldots, w_n)$  and  $\delta = (u_1, \ldots, u_m)$ . Then

$$ST(v_j) = S\left(\sum_{k=1}^n b_{kj} w_k\right)$$

$$= \sum_{k=1}^n b_{kj} S(w_k)$$

$$= \sum_{k=1}^n b_{kj} \sum_{i=1}^m a_{ik} u_i$$

$$= \sum_{i=1}^m \left(\sum_{k=1}^n a_{ik} b_{kj}\right) u_i.$$

Thus  $[ST]^{\gamma}_{\beta} = (c_{ij})$  where  $c_{ij} = \sum_{k=1}^{n} a_{ik} b_{kj}$ .

We can translate definitions and results about linear transformations into matrices.

**Definition 5.5.** For  $A \in M_{m \times n}(F)$ , we define  $L_A : F^n \to F^m$  to be the linear transformation  $L_A(x) = Ax$ . (We sometimes do not distinguish A and  $L_A$ .) The following notation are inherited from those of linear transformations

 $\ker A = \ker L_A$ ,  $\operatorname{null} A = \operatorname{null} L_A$ ,  $\operatorname{im} A = \operatorname{im} L_A$ ,  $\operatorname{rank} A = \operatorname{rank} L_A$ .

The following properties of matrix multiplication follows directly from Proposition 5.4 and Proposition 4.2 applied to  $L_A$ ,  $L_B$ ,  $L_C$ .

**Proposition 5.6** (Matrix algebra rules). For any matrices of compatible sizes and any scalar  $a \in F$ :

- (i) A(BC) = (AB)C
- (ii) A(B+C) = AB + AC
- (iii) (A+B)C = AC + BC
- (iv) a(BC) = (aB)C = B(aC)
- (iv)  $AI_n = A$ ,  $I_m A = A$

The following rank nullity theorem for matrix follows from applying Corollary 4.14 to  $L_A$ .

**Theorem 5.7.** Let  $A \in M_{m \times n}(F)$ . Then

$$n = \operatorname{rank} A + \operatorname{null} A$$
.

**Definition 5.8.** A matrix A is called **invertible** if  $L_A$  is invertible.

By Corollary 4.11, if A is invertible, then A is a square matrix. By Corollary 4.16, we have the following invertibility criterion.

Corollary 5.9. Let  $A \in M_{n \times n}(F)$ . The following are equivalent.

- (i) A is invertible.
- (ii)  $\operatorname{null} A = 0$ .
- (iii) rank A = n.
- (iv) There is  $B \in M_{n \times n}(F)$  such that AB = I.
- (v) There is  $B \in M_{n \times n}(F)$  such that BA = I.

If any of the above holds, then the matrix B in (iv) (v) are the same and unique.

**Definition 5.10.** If A is invertible, then the B in (iv), (v) above are called the **inverse** matrix of A denoted by  $A^{-1}$ .

**Corollary 5.11.** Let  $T: V \to W$  be a linear transformation,  $\beta$  be a basis of V and  $\gamma$  be a basis of W. Then T is invertible if and only if  $[T]^{\gamma}_{\beta}$  is invertible. Furthermore,

$$[T^{-1}]^{\beta}_{\gamma} = ([T]^{\gamma}_{\beta})^{-1}.$$

*Proof.* If T is invertible, then  $I = [\operatorname{id}_V]_{\beta}^{\beta} = [T^{-1}T]_{\beta}^{\beta} = [T^{-1}]_{\gamma}^{\beta}[T]_{\beta}^{\gamma}$ . Then  $[T]_{\beta}^{\gamma}$  is invertible and we have  $([T]_{\beta}^{\gamma})^{-1} = [T^{-1}]_{\gamma}^{\beta}$ .

If  $A = [T]_{\beta}^{\gamma}$  is invertible, then since  $\phi_{\gamma} T \phi_{\beta}^{-1} = L_A$ , we have  $T = \phi_{\gamma}^{-1} L_A \phi_{\beta}$ . Since the right hand side are all invertible, we have T is invertible by Lemma 4.12.

### Change of basis formula

**Lemma 5.12.** Let  $\beta = (v_1, \ldots, v_n)$  and  $\gamma = (w_1, \ldots, w_n)$  be two bases of V. Let  $P = (p_{ij}) = [\mathrm{id}]_{\beta}^{\gamma} \in M_{n \times n}(F)$ . Then  $v_j = \sum_{i=1}^m p_{ij}w_i$  for  $1 \leq j \leq n$  and for any  $v \in V$ ,  $[v]_{\gamma} = P[v]_{\beta}.$ 

*Proof.* This is a direct consequence of Theorem 4.22 with T = id. 

**Definition 5.13.** The change of coordinates matrix from  $\beta$ -coordinates to  $\gamma$ -coordinates is the matrix  $P = (p_{ij}) \in M_{n \times n}(F)$  such that  $[v]_{\gamma} = P[v]_{\beta}$  for any  $v \in V$ . In other words,  $P = [\mathrm{id}]^{\gamma}_{\beta} = ([v_1]_{\gamma}, \dots, [v_n]_{\gamma}).$ 

The matrix P is also called the **change of basis matrix** (transition matrix) from  $\gamma$  to  $\beta$  in view of the property  $v_j = \sum_{i=1}^n p_{ij} w_i$  for any  $1 \leq j \leq n$ .

Remark 5.14. We note here that if P is the change of basis matrix from  $\gamma$  to  $\beta$ , then it is the change of coordinates matrix from  $\beta$ -coordinates to  $\gamma$ -coordinates. The role of  $\beta$  and  $\gamma$  are switched when we switch our view point from change of basis to change of coordinates.

Remark 5.15. We also note difference between the definition of change of basis matrix,  $v_j = \sum_{i=1}^m p_{ij} w_i$  and the usual definition of a matrix A multiplying a vector x,  $\sum_{j=1}^{n} a_{ij}x_{j}$ . In the first case, the first index of the matrix P is summed over with  $w_i$  which is a vector, while in the second case, the second index of the matrix A is summed over with  $x_i$  which is a component of a vector. This is why some people will write  $(v_1, \ldots, v_n) = (w_1, \ldots, w_n)P$  for the change of basis matrix thought of as the "row vector"  $(w_1, \ldots, w_n)$  multiplying the matrix P in the usual way.

**Lemma 5.16.** Let  $P = [id]^{\gamma}_{\beta}$  be the change of basis matrix from  $\gamma$  to  $\beta$ . Then P is invertible and  $P^{-1} = [id]_{\gamma}^{\beta}$  is the change of basis matrix from  $\beta$  to  $\gamma$ .

*Proof.* Since id is invertible and  $id^{-1} = id$ , by Corollary 5.11 we have  $P = [id]_{\beta}^{\gamma}$  is invertible and  $P^{-1} = [id]_{\gamma}^{\beta}$ .

**Example 5.17.** Let  $\beta$  be the standard basis of  $F^3$  and  $\gamma = ((1,0,0)^t, (1,1,0)^t, (1,1,1)^t)$ ,

then the change of basis matrix from  $\beta$  to  $\gamma$  is  $[\mathrm{id}]_{\gamma}^{\beta} = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix}$ . In other words,

 $w_1 = v_1, w_2 = v_1 + v_2, w_3 = v_1 + v_2 + v_3$ . To find the inverse matrix, we have  $v_1 = w_1$ ,

$$v_2 = w_2 - v_1 = w_2 - w_1, v_3 = w_3 - v_1 - v_2 = w_3 - w_2.$$
 Then  $[id]_{\beta}^{\gamma} = \begin{pmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{pmatrix}.$ 

**Theorem 5.18.** Let  $T: V \to W$  be a linear transformation,  $\beta$ ,  $\beta'$  be bases of V and  $\gamma$ ,  $\gamma'$  be bases of W. Suppose  $A = [T]_{\beta}^{\gamma}$ ,  $B = [T]_{\beta'}^{\gamma'}$ ,  $P = [\mathrm{id}_V]_{\beta'}^{\beta}$  and  $Q = [\mathrm{id}_W]_{\gamma'}^{\gamma}$ . Then we have

$$B = Q^{-1}AP.$$

*Proof.* By Proposition 5.4, 
$$B = [T]_{\beta'}^{\gamma'} = [\operatorname{id}]_{\gamma}^{\gamma'}[T]_{\beta}^{\gamma}[\operatorname{id}]_{\beta'}^{\beta} = Q^{-1}AP$$
.

**Corollary 5.19.** Let  $T: V \to V$  be a linear transformation and  $\beta$ ,  $\beta'$  be two bases of V. Suppose  $A = [T]^{\beta}_{\beta}$ ,  $B = [T]^{\beta'}_{\beta'}$  and  $P = [\operatorname{id}]^{\beta}_{\beta'}$  is the change of basis matrix from  $\beta$  to  $\beta'$ . Then

$$B = P^{-1}AP.$$

**Definition 5.20.** Given  $A, B \in M_{n \times n}(F)$ , we say that A and B are similar if there is  $P \in M_{n \times n}(F)$  invertible such that  $B = P^{-1}AP$ .

Finding the simplest matrix which is similar to a given  $A \in M_{n \times n}(F)$  is the main reason to introduce eigenvalues and eigenvectors which will be discussed later in the notes. However, if one allows different bases of V and W as in Theorem 5.18, then we have the following theorem.

**Theorem 5.21.** Let  $T: V \to W$  be a linear transformation. Then there exists a basis  $\beta$  of V and a basis  $\gamma$  of W such that

$$[T]_{\beta}^{\gamma} = \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix},$$

where  $r = \operatorname{rank} T$ . In particular, the rank is the only invariant of a linear transformation under change of basis on both the domain and the target.

In matrix language, for any  $A \in M_{m \times n}(F)$ , then there exist invertible matrices  $P \in M_{n \times n}(F)$  and  $Q \in M_{m \times m}(F)$  such that  $Q^{-1}AP = \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix}$ .

Proof. Let  $v_1, \ldots, v_n$  be a basis of V. Then im  $T = \operatorname{span}(T(v_1), \ldots, T(v_n))$ . By Lemma 2.11,  $(T(v_1), \ldots, T(v_n))$  contains a basis of im T which after relabeling the indices, we assume to be  $T(v_1), \ldots, T(v_r)$  for  $r = \operatorname{rank} T$ . In particular,  $T(v_1), \ldots, T(v_r)$  is linearly independent. By Corollary 2.16, we can extend it to be  $\gamma = (T(v_1), \ldots, T(v_r), w_{r+1}, \ldots, w_m)$  which is a basis of W. Since  $\operatorname{span}(T(v_1), \ldots, T(v_r)) = \operatorname{im} T$ , there exist  $a_{ij} \in F$  for  $1 \leq i \leq r$  and  $r+1 \leq j \leq n$  such that  $T(v_j) = \sum_{i=1}^r a_{ij} T(v_i)$ . We define  $e_j = v_j - \sum_{i=1}^r a_{ij} v_i$  for  $r+1 \leq j \leq n$ . Apply Lemma 2.12 to  $e_{r+1}, \ldots, e_n$  one at a time, we see that  $\beta = (v_1, \ldots, v_r, e_{r+1}, \ldots, e_n)$  is a basis of V. By definition of  $e_j$ , we have  $T(e_j) = 0$  for  $r+1 \leq j \leq n$ . Then under the basis  $\beta$  and  $\gamma$ , we have  $[T]_{\beta}^{\gamma} = \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix}$ .  $\square$ 

### 5.3 Dual space

**Definition 5.22.** A linear function (linear functional/ 1-form/covector) on V is a linear transformation  $\varphi: V \to F$ . The **dual space**  $V^*$  of V is the set of all linear functions on V. In other words  $V^* = \mathcal{L}(V, F)$ . In particular,  $V^*$  is a vector space. The **natural pairing**  $\langle \cdot, \cdot \rangle: V^* \times V \to F$  is defined as  $\langle \varphi, x \rangle := \varphi(x)$  for  $x \in V, \varphi \in V^*$ .

Remark 5.23. In quantum mechanics, a linear function is also called a bra-vector denoted by  $\langle \varphi |$  while a usual vector is a ket-vector denoted by  $|\psi\rangle$ . To pair them, just write them together  $\langle \varphi | \psi \rangle$  and it becomes a "bracket" which is the combined word of bra and ket.

Remark 5.24. When we don't have an inner product on a vector space, e.g. for vector spaces over finite fields, the natural paring serves as an "inner product". This is why we write it the same way as an inner product.

**Example 5.25.** For  $V = F^n$ , every linear functional  $\varphi : F^n \to F$  is given by

$$\varphi \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = (a_1, \dots, a_n) \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = a_1 x_1 + \dots + a_n x_n.$$

for some row vector  $(a_1, \ldots, a_n) \in M_{1 \times n}(F)$ . Thus  $(F^n)^*$  the space of row vectors while  $F^n$  is the space of column vectors.

**Definition 5.26.** Given a basis  $\beta = (v_1, \dots, v_n)$  of V, the **dual basis**  $\beta^* = (\varphi_1, \dots, \varphi_n)$  of  $V^*$  is defined by  $\varphi_i(v_j) = \delta_{ij}$  for  $1 \leq i, j \leq n$ . By Lemma 4.21,  $\varphi_i$  is uniquely determined by its value at a basis.

**Proposition 5.27.** The dual basis is a basis of  $V^*$ . In particular, dim  $V^* = \dim V$ .

*Proof.* Let  $\varphi_1, \ldots, \varphi_n$  be the dual basis of  $\beta = (v_1, \ldots, v_n)$ . We first show that  $\varphi_1, \ldots, \varphi_n$  is linearly independent. Suppose  $a_1\varphi_1 + \cdots + a_n\varphi_n = 0$ . We evaluate both sides on  $v_i$  for  $1 \le i \le n$  and get  $a_1\varphi_1(v_i) + \cdots + a_n\varphi_n(v_i) = 0$ . Since  $\varphi_i(v_j) = \delta_{ij}$ , we have  $a_i = 0$  for all  $1 \le i \le n$ . Hence  $\varphi_1, \ldots, \varphi_n$  is linearly independent.

Next we show that  $\operatorname{span}(\varphi_1,\ldots,\varphi_n)=V^*$ . Let  $\varphi\in V^*$ . For any  $v\in V$  with  $[v]_{\beta}=(x_1,\ldots,x_n)^t$ , we have  $\varphi_i(v)=x_i$  for  $1\leq i\leq n$ . We write  $a_1=\varphi(v_1),\ldots,a_n=\varphi(v_n)$ . Then by Lemma 4.21,  $\varphi(v)=a_1x_1+\cdots+a_nx_n=a_1\varphi_1(v)+\cdots+a_n\varphi_n(v)=(a_1\varphi_1+\cdots+a_n\varphi_n)(v)$ . Since this is true for any  $v\in V$  we have  $\varphi=a_1\varphi_1+\cdots+a_n\varphi_n$ .  $\square$ 

Remark 5.28. From the proof above, we see that the coordinates of v under the basis  $\beta$  is given by  $\phi_{\beta}(v) = (\varphi_1(v), \dots, \varphi_n(v))^t$ .

### 5.4 Transpose of a linear transformation

**Definition 5.29.** Let  $T: V \to W$  be a linear transformation. We define the **transpose** map (dual map/pull-back)  $T^t: W^* \to V^*$  as  $T^t(\varphi) = \varphi \circ T$ . That is, for any  $\varphi \in V^*$  and  $v \in V$ ,  $(T^t(\varphi))(v) = \varphi(T(v))$  or  $\langle T^t \varphi, v \rangle = \langle \varphi, Tv \rangle$ . Since  $T^t(\varphi) = \varphi \circ T$ ,  $T^t(\varphi)$  is a linear function on V since it is a composition of linear transformations. The map  $T^t$  is linear due to the algebraic rules of composition by linear transformations in Proposition 4.2.

**Proposition 5.30.** Let  $T: V \to W$  be a linear transformation. Let  $\beta = (v_1, \ldots, v_n)$  be a basis of V and  $\gamma = (w_1, \ldots, w_m)$  be a basis of W. Let  $\beta^*$  be the dual basis of  $\beta$  and  $\gamma^*$  be the dual basis of  $\gamma$ . Then  $[T^t]_{\gamma^*}^{\beta^*} = ([T]_{\beta}^{\gamma})^t$ .

*Proof.* Let  $\gamma^* = (\psi_1, \dots, \psi_m)$  be the dual basis of  $\gamma$  and  $[T]^{\gamma}_{\beta} = (a_{ij})$ . We want to find the coordinates of  $T^t(\psi_i)$  under  $\beta^* = (\varphi_1, \dots, \varphi_n)$ , the dual basis of  $\beta$ . Let  $v \in V$  and

the coordinates of 
$$T$$
 ( $\psi_i$ ) under  $\beta = (\varphi_1, \dots, \varphi_n)$ , the dual basis of  $\beta$ . Let  $v \in V$  and  $v = x_1v_1 + \dots + x_nv_n$ . Note that  $x_i = \varphi_i(v)$ . Then we have  $T^t(\psi_i)(v) = \psi_i(T(v)) = \psi_i(\sum_{k=1}^n (\sum_{j=1}^n a_{kj}x_j)w_k) = \sum_{j=1}^n a_{ij}x_j = \sum_{j=1}^n a_{ij}\varphi_j(v) = (\sum_{j=1}^n a_{ij}\varphi_j)(v)$ . Thus  $T^t(\psi_i) = \sum_{j=1}^n a_{ij}\varphi_j$ . That is  $[T^t(\psi_i)]_{\gamma^*} = \begin{pmatrix} a_{i1} \\ \vdots \\ a_{in} \end{pmatrix}$ . Hence  $[T^t]_{\gamma^*}^{\beta^*} = ([T]_{\beta}^{\gamma})^t$ .